

Creating Linear Regression Model Using PySpark

Install PySpark

```
In [ ]: pip install pyspark
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting pyspark
  Downloading pyspark-3.2.1.tar.gz (281.4 MB)
    |████████████████████████████████████████| 281.4 MB 34 kB/s
Collecting py4j==0.10.9.3
  Downloading py4j-0.10.9.3-py2.py3-none-any.whl (198 kB)
    |████████████████████████████████████████| 198 kB 46.5 MB/s
Building wheels for collected packages: pyspark
  Building wheel for pyspark (setup.py) ... done
  Created wheel for pyspark: filename=pyspark-3.2.1-py2.py3-none-any.whl size=281853642 sha256=99657e37a6edb52a83d4b4e280e11c4e26947120a4b2dc8192749712f371238e
  Stored in directory: /root/.cache/pip/wheels/9f/f5/07/7cd8017084dce4e93e84e92efd1e1d5334db05f2e83bcef74f
Successfully built pyspark
Installing collected packages: py4j, pyspark
Successfully installed py4j-0.10.9.3 pyspark-3.2.1
```

Import Library and Creating Session

```
In [ ]: from pyspark.sql import SparkSession
```

```
In [ ]: session = SparkSession.builder.appName("exam1").master("local").getOrCreate()
```

Read Dataset

```
In [ ]: data = session.read.csv("Big Mart Sale.csv", header = True, inferSchema=True)
```

To print top 10 raw in dataset

```
In [ ]: data.show(10)
```

```
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+-----+-----+-----+-----+
+-----+-----+-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_MRP|Outlet_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|Outlet_Type|Item_Outlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
|FDA15|9.3|Low Fat|0.016047301|Dairy|249.8092|OUT049|1999|Medium|Tier 1|Supermarket Type1|3735.138|
|DRC01|5.92|Regular|0.019278216|Soft Drinks|48.2692|OUT018|2009|Medium|Tier 3|Supermarket Type2|443.4228|
```

```

|          FDN15|          17.5|          Low Fat|          0.016760075|          Meat|          141.
618|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          2097.27|
|          FDX07|          19.2|          Regular|          0.0|Fruits and Vegeta...| 182.
095|          OUT010|          1998|          null|          Tier 3|Gro
cery Store|          732.38|
|          NCD19|          8.93|          Low Fat|          0.0|          Household| 53.8
614|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          994.7052|
|          FDP36|          10.395|          Regular|          0.0|          Baking Goods| 51.4
008|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          556.6088|
|          FDO10|          13.65|          Regular|          0.012741089|          Snack Foods| 57.6
588|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          343.5528|
|          FDP10|          null|          Low Fat|          0.127469857|          Snack Foods|107.7
622|          OUT027|          1985|          Medium|          Tier 3|Superma
rket Type3|          4022.7636|
|          FDH17|          16.2|          Regular|          0.016687114|          Frozen Foods| 96.9
726|          OUT045|          2002|          null|          Tier 2|Superma
rket Type1|          1076.5986|
|          FDU28|          19.2|          Regular|          0.09444959|          Frozen Foods|187.8
214|          OUT017|          2007|          null|          Tier 2|Superma
rket Type1|          4710.535|
+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+
-----+-----+
only showing top 10 rows

```

Check Null Values in columns

```
In [ ]: from pyspark.sql.functions import isnan, when, count, col
```

```
In [ ]: data.select([count(when(isnan(c) | col(c).isNull(), c)).alias(c) for c in data.columns])

+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+
-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_MRP|Outlet_
Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|Outlet_Type|Item_O
utlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+
-----+
|          0|          1463|          0|          0|          0|          0|          0|
          0|          0|          2410|          0|          0|          0|
          0|
+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+
-----+
```

```
In [ ]: import pyspark.sql.functions as func
```

```
In [ ]: data.agg(func.percentile_approx("Item_Weight", 0.5).alias("mean")).show()

+----+
|mean|
+----+
|12.6|
+----+
```

Fill Null Values

First we replace 12.6 in place of Null values in Item_weight column because it is mean in this column

```
In [ ]: data = data.na.fill(value=12.6,subset=["Item_Weight"])
```

Second we return Medium in place of Null values in Outlet_Size Column Because Medium is the median in Outlet_Size Column

```
In [ ]: data = data.na.fill(value="Medium",subset=["Outlet_Size"])
```

```
In [ ]: data.show()
```

```
+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+
|Item_Identifier|Item_Weight|Item_Fat_Content|Item_Visibility|Item_Type|Item_
MRP|Outlet_Identifier|Outlet_Establishment_Year|Outlet_Size|Outlet_Location_Type|Outlet_Type|Item_Outlet_Sales|
+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+
|          FDA15|          9.3|          Low Fat|          0.016047301|          Dairy|249.8
092|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          3735.138|
|          DRC01|          5.92|          Regular|          0.019278216|          Soft Drinks| 48.2
692|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          443.4228|
|          FDN15|          17.5|          Low Fat|          0.016760075|          Meat| 141.
618|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          2097.27|
|          FDX07|          19.2|          Regular|          0.0|Fruits and Vegeta...| 182.
095|          OUT010|          1998|          Medium|          Tier 3|Gro
cery Store|          732.38|
|          NCD19|          8.93|          Low Fat|          0.0|          Household| 53.8
614|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          994.7052|
|          FDP36|          10.395|          Regular|          0.0|          Baking Goods| 51.4
008|          OUT018|          2009|          Medium|          Tier 3|Superma
rket Type2|          556.6088|
|          FDO10|          13.65|          Regular|          0.012741089|          Snack Foods| 57.6
588|          OUT013|          1987|          High|          Tier 3|Superma
rket Type1|          343.5528|
|          FDP10|          12.6|          Low Fat|          0.127469857|          Snack Foods|107.7
622|          OUT027|          1985|          Medium|          Tier 3|Superma
rket Type3|          4022.7636|
|          FDH17|          16.2|          Regular|          0.016687114|          Frozen Foods| 96.9
726|          OUT045|          2002|          Medium|          Tier 2|Superma
rket Type1|          1076.5986|
|          FDU28|          19.2|          Regular|          0.09444959|          Frozen Foods|187.8
214|          OUT017|          2007|          Medium|          Tier 2|Superma
rket Type1|          4710.535|
|          FDY07|          11.8|          Low Fat|          0.0|Fruits and Vegeta...| 45.5
402|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          1516.0266|
|          FDA03|          18.5|          Regular|          0.045463773|          Dairy|144.1
102|          OUT046|          1997|          Small|          Tier 1|Superma
rket Type1|          2187.153|
|          FDX32|          15.1|          Regular|          0.1000135|Fruits and Vegeta...|145.4
786|          OUT049|          1999|          Medium|          Tier 1|Superma
rket Type1|          1589.2646|
|          FDS46|          17.6|          Regular|          0.047257328|          Snack Foods|119.6
782|          OUT046|          1997|          Small|          Tier 1|Superma
```

```

rket Type1|      2145.2076|
|      FDF32|      16.35|      Low Fat|      0.0680243|Fruits and Vegeta...|196.4
426|      OUT013|      1987|      High|      Tier 3|Superma
rket Type1|      1977.426|
|      FDP49|      9.0|      Regular|      0.069088961|      Breakfast| 56.3
614|      OUT046|      1997|      Small|      Tier 1|Superma
rket Type1|      1547.3192|
|      NCB42|      11.8|      Low Fat|      0.008596051|      Health and Hygiene|115.3
492|      OUT018|      2009|      Medium|      Tier 3|Superma
rket Type2|      1621.8888|
|      FDP49|      9.0|      Regular|      0.069196376|      Breakfast| 54.3
614|      OUT049|      1999|      Medium|      Tier 1|Superma
rket Type1|      718.3982|
|      DRI11|      12.6|      Low Fat|      0.034237682|      Hard Drinks|113.2
834|      OUT027|      1985|      Medium|      Tier 3|Superma
rket Type3|      2303.668|
|      FDU02|      13.35|      Low Fat|      0.10249212|      Dairy|230.5
352|      OUT035|      2004|      Small|      Tier 2|Superma
rket Type1|      2748.4224|
+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+
-----+-----+
only showing top 20 rows

```

A Simple Exploratory Of Dataset

To print all columns name

```

In [ ]: data.columns

Out[ ]: ['Item_Identifier',
        'Item_Weight',
        'Item_Fat_Content',
        'Item_Visibility',
        'Item_Type',
        'Item_MRP',
        'Outlet_Identifier',
        'Outlet_Establishment_Year',
        'Outlet_Size',
        'Outlet_Location_Type',
        'Outlet_Type',
        'Item_Outlet_Sales']

```

To count total numbers of raws in dataset

```

In [ ]: data.count()

Out[ ]: 8523

```

To print the schema of dataset, Spark schema is the structure of the DataFrame or Dataset, which is a collection of StructField that define the column name(String), column type (DataType), nullable column (Boolean) and metadata (MetaData)

```

In [ ]: data.printSchema()

root
 |-- Item_Identifier: string (nullable = true)
 |-- Item_Weight: double (nullable = false)
 |-- Item_Fat_Content: string (nullable = true)
 |-- Item_Visibility: double (nullable = true)
 |-- Item_Type: string (nullable = true)

```

```

|-- Item_MRP: double (nullable = true)
|-- Outlet_Identifier: string (nullable= true)
|-- Outlet_Establishment_Year: integer (nullable = true)
|-- Outlet_Size: string (nullable = false)
|-- Outlet_Location_Type: string (nullable = true)
|-- Outlet_Type: string (nullable = true)
|-- Item_Outlet_Sales: double (nullable = true)

```

To know data type of each columns

```
In [ ]: data.dtypes
```

```

Out[ ]: [('Item_Identifier', 'string'),
('Item_Weight', 'double'),
('Item_Fat_Content', 'string'),
('Item_Visibility', 'double'),
('Item_Type', 'string'),
('Item_MRP', 'double'),
('Outlet_Identifier', 'string'),
('Outlet_Establishment_Year', 'int'),
('Outlet_Size', 'string'),
('Outlet_Location_Type', 'string'),
('Outlet_Type', 'string'),
('Item_Outlet_Sales', 'double')]

```

Data Preprocessing

Here we convert the data into machine readable form

VectorAssembler :- It is feature transformer that combine multiple columns into a single vector column.

StringIndexer :- It is use for mapping a string column to a index column that will be treated as a categorical column by spark.

OneHotEncoder :- It is an important technique for converting categorical attributes into a numeric vector

```
In [ ]: from pyspark.ml.feature import VectorAssembler, StringIndexer, OneHotEncoder
```

```
In [ ]: str_index = StringIndexer(inputCols = ['Item_Identifier','Item_Fat_Content','Item_Type',
```

```
In [ ]: one_hot = OneHotEncoder(inputCols =['Item_Identifier1','Item_Fat_Content1','Item_Type1',
```

```
In [ ]: vector_ass = VectorAssembler(inputCols = ['Item_Weight','Item_Fat_Content2','Item_Visibi
```

Import Linear Regression and Create Model

```
In [ ]: from pyspark.ml.regression import LinearRegression
```

```
In [ ]: linear = LinearRegression(featuresCol="allfeatures", labelCol="Item_Outlet_Sales")
```

Create Pipeline for ML Model

```

In [ ]: from pyspark.ml import Pipeline
mypipeline = Pipeline(stages = [str_index, one_hot, vector_ass, linear])

```

Item and Outlet Information																	
Item_Identifier	Item_Weight	Item_Fat_Content	Item_Visibility	Item_Type	Item_MRP	Outlet_Identifier	Outlet_Establishment_Year	Outlet_Size	Outlet_Location_Type	Outlet_Type	Item_Outlet_Sales	Item_Outlet_Sales					
Item_Identifier1	Item_Fat_Content1	Item_Type1	Outlet_Identifier1	Outlet_Establishment_Year1	Outlet_Size1	Outlet_Location_Type1	Outlet_Type1	Item_Identifier2	Item_Fat_Content2	Item_Type2	Outlet_Identifier2	Outlet_Establishment_Year2	Outlet_Size2	Outlet_Location_Type2	Outlet_Type2	allfeatures	prediction
1	DRA12	11.6	Low Fat	0.0	Soft Drinks	141.6154	OUT045	2002	Medium	Tier 2	Supermarket Type	3829.0158	1051.0	0.0	8.0	7.0	
	(4, [0], [1.0])	(15, [8], [1.0])	(9, [7], [1.0])	(8, [7], [1.0])	(2, [0], [1.0])	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 1, 14, 21, 22, ...])	2277.372227927723								
2	DRA12	11.6	Low Fat	0.041112694	Soft Drinks	142.0154	OUT018	2009	Medium	Tier 3	Supermarket Type	850.8924	1051.0	0.0	8.0	5.0	
	(4, [0], [1.0])	(15, [8], [1.0])	(9, [5], [1.0])	(8, [5], [1.0])	(2, [0], [1.0])	(2, [0], [1.0])	(3, [], [])	(29, [0, 1, 5, 14, 21, ...])	1937.56029983537								
	DRA12	11.6	Low Fat	0.068535039	Soft Drinks	143.0154	OUT010	1998	Medium	Tier 3	Grocery Store	283.6308	1051.0	0.0	8.0	8.0	
	(4, [0], [1.0])	(15, [8], [1.0])	(9, [8], [1.0])	(8, [], [])	(2, [0], [1.0])	(2, [0], [1.0])	(3, [1], [1.0])	(29, [0, 1, 5, 14, 21, ...])	327.38010298800816								
1	DRA24	19.35	Regular	0.039920687	Soft Drinks	163.3868	OUT035	2004	Small	Tier 2	Supermarket Type	3439.5228	322.0	1.0	8.0	1.0	
	(4, [0], [1.0])	(15, [8], [1.0])	(9, [8], [1.0])	(8, [], [])	(2, [0], [1.0])	(2, [0], [1.0])	(3, [1], [1.0])	(29, [0, 1, 5, 14, 21, ...])	1553.322								

0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])	(8, [2], [1.0])	(2, [1], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 2, 5, 14, 21, ...]	2671.7706482761328	
	DRA24	19.35	Regular	0.040154087	Soft Drinks 164.6868
	OUT017		2007	Medium	Tier 2 Supermarket Type
1	1146.5076	322.0	1.0	8.0	2.0
	3.0	0.0	1.0	0.0	(1553, [322], [1.0])
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [2], [1.0])	(8, [3], [1.0])	(2, [0], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 2, 5, 14, 21, ...]	2691.662378943857	
	DRA59	8.27	Regular	0.0	Soft Drinks 183.2924
	OUT017		2007	Medium	Tier 2 Supermarket Type
1	2406.2012	97.0	1.0	8.0	2.0
	3.0	0.0	1.0	0.0	(1553, [97], [1.0])
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [2], [1.0])	(8, [3], [1.0])	(2, [0], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 2, 14, 21, 22, ...]	3009.0974065421287	
	DRA59	12.6	Regular	0.127308434	Soft Drinks 186.6924
	OUT027		1985	Medium	Tier 3 Supermarket Type
3	7033.5112	97.0	1.0	8.0	4.0
	0.0	0.0	0.0	2.0	(1553, [97], [1.0])
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [4], [1.0])	(8, [0], [1.0])	(2, [0], [1.0])
	(2, [0], [1.0])	(3, [2], [1.0])	(29, [0, 2, 5, 14, 21, ...]	4393.2781578674	
	DRB01	7.39	Low Fat	0.082367244	Soft Drinks 187.753
	OUT049		1999	Medium	Tier 1 Supermarket Type
1	1518.024	1336.0	0.0	8.0	3.0
	4.0	0.0	2.0	0.0	(1553, [1336], [1.0])
0))	(4, [0], [1.0])	(15, [8], [1.0])	(9, [3], [1.0])	(8, [4], [1.0])	(2, [0], [1.0])
	(2, [], [])	(3, [0], [1.0])	(29, [0, 1, 5, 14, 21, ...]	2996.614390466222	
	DRB13	6.115	Regular	0.007043008	Soft Drinks 190.353
	OUT035		2004	Small	Tier 2 Supermarket Type
1	569.259	1052.0	1.0	8.0	1.0
	2.0	1.0	1.0	0.0	(1553, [1052], [1.0])
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])	(8, [2], [1.0])	(2, [1], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 2, 5, 14, 21, ...]	3120.027487849694	
	DRB13	6.115	Regular	0.01179078	Soft Drinks 189.053
	OUT010		1998	Medium	Tier 3 Grocery Stor
e	948.765	1052.0	1.0	8.0	8.0
	8.0	0.0	0.0	1.0	(1553, [1052], [1.0])
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [8], [1.0])	(8, [], [])	(2, [0], [1.0])
	(2, [0], [1.0])	(3, [1], [1.0])	(29, [0, 2, 5, 14, 21, ...]	1146.3741600280819	
	DRB25	12.3	Low Fat	0.069446588	Soft Drinks 106.3938
	OUT035		2004	Small	Tier 2 Supermarket Type
1	857.5504	323.0	0.0	8.0	1.0
	2.0	1.0	1.0	0.0	(1553, [323], [1.0])
0))	(4, [0], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])	(8, [2], [1.0])	(2, [1], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 1, 5, 14, 21, ...]	1711.1967642199536	
	DRB48	12.6	Regular	0.024733134	Soft Drinks 40.2822
	OUT027		1985	Medium	Tier 3 Supermarket Type
3	1296.3126	672.0	1.0	8.0	4.0
	0.0	0.0	0.0	2.0	(1553, [672], [1.0])
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [4], [1.0])	(8, [0], [1.0])	(2, [0], [1.0])
	(2, [0], [1.0])	(3, [2], [1.0])	(29, [0, 2, 5, 14, 21, ...]	2156.0649493401916	
	DRB48	16.75	Regular	0.024848788	Soft Drinks 39.9822
	OUT035		2004	Small	Tier 2 Supermarket Type
1	746.3618	672.0	1.0	8.0	1.0
	2.0	1.0	1.0	0.0	(1553, [672], [1.0])
0))	(4, [1], [1.0])	(15, [8], [1.0])	(9, [1], [1.0])	(8, [2], [1.0])	(2, [1], [1.0])
	(2, [1], [1.0])	(3, [0], [1.0])	(29, [0, 2, 5, 14, 21, ...]	769.7994983213209	

```

0])|      (4,[1],[1.0])|(15,[8],[1.0])|      (9,[5],[1.0])|      (8,[5],[1.0])|(2,
[0],[1.0])|      (2,[0],[1.0])|      (3,[],[])|(29,[0,2,5,14,21,...| 582.7858526373045|
|      DRC01|      5.92|      Regular|      0.019308607|Soft Drinks| 49.0692|
      OUT017|      2007|      Medium|      Tier 2|Supermarket Type
1|      1478.076|      673.0|      1.0|      8.0|      2.0|
      3.0|      0.0|      1.0|      0.0|(1553,[673],[1.
0])|      (4,[1],[1.0])|(15,[8],[1.0])|      (9,[2],[1.0])|      (8,[3],[1.0])|(2,
[0],[1.0])|      (2,[1],[1.0])|(3,[0],[1.0])|(29,[0,2,5,14,21,...| 929.3680572902323|
|      DRC12|      17.85|      Low Fat|      0.03781972|Soft Drinks|191.6188|
      OUT035|      2004|      Small|      Tier 2|Supermarket Type
1|      2475.4444|      1498.0|      0.0|      8.0|      1.0|
      2.0|      1.0|      1.0|      0.0|(1553,[1498],[1.
0])|      (4,[0],[1.0])|(15,[8],[1.0])|      (9,[1],[1.0])|      (8,[2],[1.0])|(2,
[1],[1.0])|      (2,[1],[1.0])|(3,[0],[1.0])|(29,[0,1,5,14,21,...| 3030.7345765579985|
|      DRC12|      17.85|      Low Fat|      0.037826873|Soft Drinks|189.7188|
      OUT046|      1997|      Small|      Tier 1|Supermarket Type
1|      2285.0256|      1498.0|      0.0|      8.0|      6.0|
      6.0|      1.0|      2.0|      0.0|(1553,[1498],[1.
0])|      (4,[0],[1.0])|(15,[8],[1.0])|      (9,[6],[1.0])|      (8,[6],[1.0])|(2,
[1],[1.0])|      (2,[],[])|(3,[0],[1.0])|(29,[0,1,5,14,21,...| 3023.129033573159|
|      DRC12|      17.85|      Low Fat|      0.038040837|Soft Drinks|189.1188|
      OUT017|      2007|      Medium|      Tier 2|Supermarket Type
1|      3237.1196|      1498.0|      0.0|      8.0|      2.0|
      3.0|      0.0|      1.0|      0.0|(1553,[1498],[1.
0])|      (4,[0],[1.0])|(15,[8],[1.0])|      (9,[2],[1.0])|      (8,[3],[1.0])|(2,
[0],[1.0])|      (2,[1],[1.0])|(3,[0],[1.0])|(29,[0,1,5,14,21,...| 2991.8000936798494|
|      DRC13|      8.26|      Regular|      0.032573725|Soft Drinks| 125.073|
      OUT018|      2009|      Medium|      Tier 3|Supermarket Type
2|      985.384|      1337.0|      1.0|      8.0|      5.0|
      5.0|      0.0|      0.0|      3.0|(1553,[1337],[1.
0])|      (4,[1],[1.0])|(15,[8],[1.0])|      (9,[5],[1.0])|      (8,[5],[1.0])|(2,
[0],[1.0])|      (2,[0],[1.0])|      (3,[],[])|(29,[0,2,5,14,21,...| 1764.2353427575733|
+-----+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+-----+
-+-----+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+-----+
---+-----+-----+-----+-----+-----+-----+-----+-----+
-----+-----+-----+-----+-----+-----+-----+-----+
only showing top 20 rows

```

Evaluate Model

```
In [ ]: from pyspark.ml.evaluation import RegressionEvaluator
```

```
In [ ]: errors = ["r2", "rmse", "mse", "mae"]
name = ["R-Square or Accuracy", "Root Mean Square Error", "Mean Square Error", "Mean Abs

for i in range(len(errors)):
    eval = RegressionEvaluator(predictionCol="prediction", labelCol='Item_Outlet_Sales', m
    print("The {} of Model is {}".format(name[i],eval.evaluate(result)))
```

```

The R-Square or Accuracy of Model is 0.5609324399455548
The Root Mean Square Error of Model is 1146.154277794764
The Mean Square Error of Model is 1313669.6285072374
The Mean Absolute Error of Model is 854.720185337692

```

```
In [ ]:
```